on said feature frame matrix.

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4	(a) selecting a frame cluster in said input video sequence which
5	corresponds to a most static one of said video segments;
6	(b) computing a content value is said selected frame cluster;
7	(c) using said computed content value to cluster remaining frames
8	in said input video sequence.
1	2. The method of claim 1, wherein in said (a) said frame cluster is
11 B B 2	selected using a refined feature space representation of said input video sequence.
- " الله	The same sequences of the second sequences of the second sequences.
1/	3. The method of claim 1, wherein in said (a) each of said plurality of
2	frames is transformed into a histogram vector indicative of a spatial distribution of
3	colors in said each of said plurality of frames.
1	4. The method of claim 3, wherein in said (a) each of said plurality of
2	frames is divided into a plurality of blocks, each of said plurality of blocks being
3	represented by a histogram in a color space indicative of a distribution of colors
4	within each of said plurality of blocks.
1	5. The method of claim 3, wherein each of said plurality of frames is
2	divided into a plurality of blocks and each said histogram vector comprises a plurality
3	of histograms in a color space, each of said plurality of histograms corresponding to
4	one of said plurality of blocks.
1	6. The method of claim $2$ , wherein said refined feature space
2	representation is obtained using a singular value decomposition of said input video
3	sequence.
1	7. The method of claim 6 wherein said singular value decomposition is
2	performed using frames selected with a fixed interval from said input video sequence.
1	8. The method of claim 7, wherein said selected frames are arranged into
2	a feature frame matrix, and wherein said singular value decomposition is performed

1	9. The method of claim 6, wherein said singular value decomposition
2	produces a matrix, each column of said matrix representing a frame in a refine
3	feature space corresponding to a frame in said input video sequence.
1	10. The method of claim 1, further comprising (d) using said clustered
2	frames to output a motion video representative of a summary of said input video
3	sequence.
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1	11. The method of claim 1, further comprising (d) outputting a plurality of
2	keyframes, each of said plurality of keyframes representative of said clustered frames
1	12. The method of claim 2, wherein said selecting comprises locating
2	cluster closest to an origin of said refined feature space.
1	13. The method of claim 2, wherein said (c) comprises:
2	(c)(1) sorting a plurality of vectors in said refined feature space i
3	ascending order according to a distance of each of said vector
4	to an origin of said refined feature space representation;
5	(c)(2) selecting a vector among said sorted vectors which is closest t
6	an origin of said refined feature space representation an
7	including said selected vector into a first cluster;
8	(c)(3) clustering said plurality of sorted vectors in said refined featur
9	into a plurality of clusters according to a distance between eac
10	of said plurality of sorted vectors and vectors in each of sai
11	plurality of clusters and an amount of information in each of
12	said plurality of clusters.
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1	14. The method of claim 13, wherein in said (c)(3) said plurality of sorte
2	vectors are clustered into said plurality of clusters such that said amount of
3	information in each of said plurality of clusters does not exceed an amount of
4	information in said first cluster.

1	15.	The n	nethod of claim 13, wherein said first cluster is composed of
2	frames based	on a d	listance variation between said frames and an average distance
3	between fram	es in sa	id first cluster.
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1	16.	The n	nethod of claim 13, wherein each of said plurality of clusters is
2	composed of	frame	s based on a distance variation between said frames and an
3	average distar	nce bety	ween frames in said each of said plurality of clusters.
1	17.	A met	thod for summarizing a content of an input video sequence, said
2	method comp	rising:	
3	•	(a)	selecting frames from said input video sequence, said selected
4			frames being taken at a fixed interval;
5		(b)	creating a feature frame matrix using said selected frames;
6		(c)	performing a singular value decomposition on said feature
7			frame matrix to obtain a matrix representing said video
8			sequence in a refined feature space;
9		(d)	selecting a cluster in said refined feature space corresponding
10			to a most static video segment;
11		(e)	computing a content value corresponding to said selected
12			cluster;
13		(f)	using said computed content value to cluster frames in said
14			input video sequence.
1	18.		thod for segmenting an input video sequence, said input video
2	sequence co	mprisin	g a plurality of frames, said plurality of frames being grouped
3	into a plurali	ty of vi	deo shots, said method comprising:
4		(a)	computing a similarity between each of said plurality of frames
5			and a frame preceding said each of said plurality of frames in
6			time;
7		(b)	segmenting said input video sequence into said plurality of
8			video shots according to said computed similarity.

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1	19. The method of claim 18, wherein said similarity is calculated using
2	refined feature space representation of said input video sequence.
1	20. The method of claim 19, wherein said refined feature space
2	representation is created using a singular value decomposition of said input video
3	sequence.
1	21. The method of claim 20, wherein said singular value decomposition i
2	performed using frames selected with a fixed interval from said input video sequence
1	22. The method of claim 21, wherein said selected frames are arranged
2	into a feature frame matrix, and wherein said singular value decomposition i
3	performed on said feature frame matrix.
1	23. The method of claim 22, wherein said performed singular value
2	decomposition produces a matrix, each column of said produced matrix comprising
3	frame in said refined feature space representing a frame in said input video sequence
1	24. The method of claim 18, further comprising (c) extracting feature
<b>2</b>	from each of said plurality of video shots.
3	25. A method for determining a similarity between a first and a second
4	frame in an input video sequence, said method comprising:
5	(a) calculating a refined feature space representation of said inpu
6	video sequence;
7	(b) using said calculated representation to compute said similarity
8	between said first and said second frames.

representation is calculated using a singular value decomposition.

The method of claim 25, wherein in said (a) said refined feature space

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video sequence.

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1	27.	The m	nethod of	f claim 18	, who	er <b>f</b> ein in s	said (b)	said o	comp	outed	similarity
2	is compared	to at lea	ast a first	t threshold	d sim	larity au	nd a sec	cond t	hresl	hold	similarity,
3	and said inpu	t video	sequence	is segme	nted	according	g to a re	sult o	f saic	d com	parison.
1	28.										computed
2	similarity is b	pelow a	first thre	shold sim	filarit	y, said e	ach of s	said pl	lurali	ity of	frames is
3	put into a one	of said	plurality	of wideo	shots	containi	ng said	prece	dent	in tir	ne frame.
	20	m.		( 1)/					<i>(</i> 4.)		
4	29.		•	<b>V</b> IN	1						computed
5	similarity is a	above a	second t	hreshold s	imila	arity, said	d each o	of said	plur	ality	of frames
6	is designated	as a sho	ot bounda	ary	,						
7	20	Œ1	.1 1		10	, .		• •	<i>a</i> >		. 1
7	30.			1							computed
8	similarity is	betweer	ı a first	threshold	simi	larity an	d a sec	cond t	hresh	iold s	similarity,
9	said each of	said plu	rality of	frames is	put i	nto a on	e of sai	d plur	ality	of vi	deo shots
10	according to	a furthe	r analysi	s perform	ed us	sing addi	tional f	rames	fron	n saic	d plurality
11	of frames.			1							
				1		1					
B	31	A con	nputer-re	adable me	ediun	i contain	ing a pr	ogran	ı for	sumr	narizing a
2	content of an	input v	rideo seq	uence, sai	id inp	out video	sequen	ice coi	mpri	sing a	a plurality
3/	of frames, sa	id plura	lity of fr	ames beir	ng gr	ouped in	to a plu	ırality	of v	ideo	segments,
4	said program	compri	sing:								
5		(a)	selectin	ng a fram	e clų	ter in s	aid inp	ut vid	eo s	equer	nce which
6				onds to a		'				-	
7		(b)	comput	ting conte	nt va	lue in sai	d select	ed fra	me c	luste	r:
8		(c)	-	•	- 1						ng frames
9		(-)	_	input vide	1			21400	101		
,			iii saiu	mput vide	T sec	<sub>1</sub> ucnce.					
1	32.	The c	omputer-	readable	medi	um of cl	aim 31	, wher	ein i	in sai	d (a) said

frame cluster is selected using a refined feature space representation of said input

- 1 33. The computer-readable medium of claim 31, wherein in said (a) each of said plurality of frames is transformed into a histogram vector indicative of a spatial distribution of colors in said each of said plurality of frames.
  - 34. The computer-readable medium of claim 33, wherein in said (a) each of said plurality of frames is divided into a plurality of blocks, each of said plurality of blocks being represented by a histogram in a color space indicative of a distribution of colors within each of said plurality of blocks.
- The computer-readable medium of claim 33, wherein each of said plurality of frames is divided into a plurality of blocks and each said histogram vector comprises a plurality of histograms in a color space, each of said plurality of histograms corresponding to one of said plurality of blocks.
  - 36. The computer-readable medium of claim 32, wherein said refined feature space representation is obtained using a singular value decomposition of said input video sequence.
  - 37. The computer-readable medium of claim 36, wherein said singular value decomposition is performed using frames selected with a fixed interval from said input video sequence.
  - 38. The computer-readable medium of claim 37, wherein said selected frames are arranged into a feature frame matrix, and wherein said singular value decomposition is performed on said feature frame matrix.
  - 39. The computer-readable medium of claim 33, wherein said singular value decomposition produces a matrix, each column of said matrix representing a frame in a refined feature space corresponding to a frame in said input video sequence.

1	40.	The computer-readable medium of claim 31, further comprising (d)
2	using said cl	ustered frames to output a video representative of a summary of said
3	input video so	equence.
1	41.	The computer-readable medium of claim 31, further comprising (d)
2	outputting a p	lurality of keyframes, each of said plurality of keyframes representative
3	of said cluste	red frames.
1	42.	The computer-readable medium of claim 32, wherein said selecting
2	comprises lo	cating a cluster closest to an origin of said refined feature space.
1	43.	The computer-readable medium of claim 32, wherein said (c)
2	comprises:	
3		(1) sorting a plurality of vectors in said refined feature space in
4		ascending order according to a distance of each of said vectors
5		to an origin of said refined feature space;
6		(2) selecting a vector among said sorted vectors which is closest to
7		an origin of said refined feature space and including said
8		selected vector into a first cluster;
9		(3) clustering said plurality of sorted vectors in said refined feature
10		into a plurality of clusters according to a distance between each
11		of said plurality of sorted vectors and each of said plurality of
12		clusters and an amount of information in each of said plurality
13		of clusters.
1	44.	The computer-readable medium of claim 38, wherein in said (3) said
2	plurality of s	orted vectors are clustered into said plurality of clusters such that said
3	amount of in	Formation in each of said plurality of clusters does not exceed an amount
4	of informatio	n in said first cluster.

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1	45.	The c	omputer-readable medium of claim 38, wherein said first cluster
2	is composed	of fran	nes based on a distance variation between said frames and said
3	first cluster.		
1	46.	The c	computer-readable medium of claim 38, wherein each of said
2	plurality of cl	usters i	s composed of frames based on a distance variation between said
3	frames and sa	id each	of said plurality of clusters.
1	47.	A con	nputer-readable medium containing a program for summarizing a
2	content of an	input v	ideo sequence, said program comprising:
3		(a)	selecting frames with a fixed interval from said input video
4			sequence;
5		(b)	creating a feature frame matrix using said selected frames;
6		(c)	performing a singular value decomposition on said feature
7			frame matrix to obtain matrix representing said video sequence
8			in refined feature space;
9		(d)	selecting a cluster in said refined feature space corresponding
10			to a most static video segment;
11		(e)	computing a content value corresponding to said selected
12			cluster;
13		(f)	using said computed content value to cluster frames in said
14			input video sequence.
1	48.	A con	nputer-readable medium containing a program for segmenting an
2	input video se	equence	e, said input video sequence comprising a plurality of frames, said
3	plurality of	frames	being grouped into a plurality of video shots, said program
4	comprising:		( <i>V</i> //
5		(a)	computing a similarity between each of said plurality of frames
6			and a subsequent in time frame;
7		(b)	segmenting said input video sequence into a plurality of shots
8			according to said computed similarity.
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- 1 49. The computer-readable medium of claim 18, wherein said similarity is 2 calculated using a refined feature space representation of said input video sequence.
- 1 50. The computer-readable medium of claim 19, wherein said refined 2 feature space representation is created using a singular value decomposition of said 3 input video sequence.
- 1 51. The computer-readable medium of claim 20, wherein said singular 2 value decomposition is performed using frames selected with a fixed interval from 3 said input video sequence.
- The computer-readable medium of claim 21, wherein said selected frames are arranged into a feature frame matrix, and wherein said singular value decomposition is performed on said feature frame matrix.
- The computer-readable medium of claim 22, wherein said performed singular value decomposition produces a matrix, each column of said produced matrix comprising a frame in said refined feature space representing a frame in said input video sequence.
  - 54. The computer-readable medium of claim 18, wherein said program further comprises (c) extracting features from each of said plurality of video shots.
- 55. A computer-readable medium containing a program for determining a similarity between a first and a second frames in an input video sequence, said program comprising:
- 6 (a) calculating a refined feature space representation of said input video sequence; and
- 8 (b) using said calculated representation to compute said similarity between said first and said second frames.

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- 10 56. The computer-readable medium of claim 25, wherein in said (a) said 11 refined feature space representation is calculated using a singular value 12 decomposition.
  - 57. The computer-readable medium of claim 18, wherein in said (b) said computed similarity is compared to at least two threshold similarities, and said input video sequence is segmented according to a result of said comparison.
- The computer-readable medium of claim 48, wherein if in said (b) said computed similarity is below a first threshold similarity, said each of said plurality of frames is put into a one of said plurality of video shots containing said precedent in time frame.
  - 59. The computer-readable medium of claim 48, wherein if in said (b) said computed similarity is above a second threshold similarity, said each of said plurality of frames is designated as a shot boundary.
  - 60. The computer-readable medium of claim 48, wherein if in said (b) said computed similarity is between a first threshold similarity and a second threshold similarity, said each of said plurality of frames is put into a one of said plurality of video shots according to a further analysis performed using additional frames from said plurality of frames.
- 1 61. The method of claim 18, further comprising (c) extracting features 2 from each of said plurality of video shots and using said extracted features to index 3 said plurality of video shots.
- 1 62. The method of claim 61, wherein said extracted features are features of 2 a video frame representative of said each of said plurality of video shots.
- 1 63. The computer-readable medium of claim 48, wherein said program 2 further comprises (c) extracting features from each of said plurality of video shots and 3 using said extracted features to index said plurality of video shots.

- 1 64. The method of claim 63, wherein said extracted features are features of a video frame representative of said each of said plurality of video shots.
- 1 65. A method of calculating a degree of visual changes in a video shot, 2 said video shot comprising a plurality of frames, said method comprising:
- performing a singular value decomposition on said plurality of frames,
  wherein said singular value decomposition produces a matrix, each
  column of said matrix representing a frame in a refined feature space
  corresponding to a frame in said plurality of frames;
- 7 (b) using said matrix to calculate said degree of visual changes in said video shot.
- 1 66. The method of claim 65, wherein said (b) comprises calculating said 2 degree of visual changes in said video shot as a sum  $\sqrt{\sum_{i=1}^{rank(A)} V_{ij}^2}$ , wherein  $v_{ij}$  are
- 3 elements of said matrix.
- 1 67. A computer-readable medium containing a program for calculating a 2 degree of visual changes in a video shot, said video shot comprising a plurality of 3 frames, said program comprising:
- performing a singular value decomposition on said plurality of frames,
  wherein said singular value decomposition produces a matrix, each
  column of said matrix representing a frame in a refined feature space
  corresponding to a frame in said plurality of frames;
- 8 (b) using said matrix to calculate said degree of visual changes in said video shot.
- 1 68. The computer-readable medium of claim 67, wherein said (b) 2 comprises calculating said degree of visual changes in said video shot as a sum
- 3  $\sqrt{\sum_{j=1}^{\text{rank(A)}} V_{ij}^2}$ , wherein  $v_{ij}$  are elements of said matrix.

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- 1 69. A method of calculating an evenness of color distributions in a video 2 shot, said video shot comprising a plurality of frames, said method comprising: 3 (a) performing a singular value decomposition on said plurality of frames, 4 wherein said singular value decomposition produces a matrix, each 5 column of said matrix representing a frame in a refined feature space 6 corresponding to a frame in said plurality of frames;
  - (b) using said matrix to calculate said evenness of color distribution in said video shot.
- The method of claim 69, wherein said (b) comprises calculating said evenness of color distribution in said video shot as a sum  $\sqrt{\sum_{j=1}^{rank(A)} \sigma_j^2 V_{ij}^2}$ , wherein said  $v_{ij}$  are elements of said matrix and said  $\sigma_j$  are singular values obtained in said singular value decomposition.
  - 71. A computer-readable medium containing a program for calculating an evenness of color distributions in a video shot, said video shot comprising a plurality of frames, said method comprising:
    - (a) performing a singular value decomposition on said plurality of frames, wherein said singular value decomposition produces a matrix, each column of said matrix representing a frame in a refined feature space corresponding to a frame in said plurality of frames;
    - (b) using said matrix to calculate said evenness of color distribution in said video shot.
- The computer readable medium of claim 71, wherein said (b) comprises calculating said evenness of color distribution in said video shot as a sum  $\sqrt{\sum_{j=1}^{rank(A)} \sigma_{j}^{2} V_{ij}^{2}}, \text{ wherein said } v_{ij} \text{ are elements of said matrix and said } \sigma_{j} \text{ are}$
- 4 singular values obtained in said singular value decomposition.